

# On Robust Two-Block Problems<sup>1</sup>

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## Abstract

In this paper we consider the following robust two-block problem that arises in estimation and in full-information control: minimize the worst-case  $H^\infty$  norm of a two-block transfer matrix whose elements contain  $H^\infty$ -norm-bounded modeling errors. We show that, when the underlying systems are single-input/single-output, and if the modeling errors are “small enough”, then the robust two-block problem can be solved by solving a one-dimensional family of appropriately-weighted “modeling-error-free” two-block problems. We also study the consequences of this result to a robust tracking problem, where the optimal solution can be explicitly found.

## 1 Introduction

In this paper we study the robust two-block problem

$$\inf_{Q(\cdot) \in H^\infty} \sup_{(\Delta P_1(\cdot), \Delta P_2(\cdot)) \in B_\delta^\infty} \|T_Q(P_1 + \delta P_1, P_2 + \delta P_1)\|_\infty, \quad (1)$$

where

$$T_Q(P_1 + \delta P_1, P_2 + \delta P_1) =$$

$$\begin{bmatrix} P_1(z) + \Delta P_1(z) + (P_2(z) + \Delta P_2(z)) Q(z) \\ Q(z) \end{bmatrix},$$

and

$$B_\delta^\infty = \{(\Delta P_1(\cdot), \Delta P_2(\cdot)), \|[\Delta P_1(\cdot) \quad \Delta P_2(\cdot)]\|_\infty \leq \delta\},$$

with  $Q(\cdot)$ ,  $P_1(\cdot)$ ,  $P_2(\cdot)$ ,  $\Delta P_1(\cdot)$  and  $\Delta P_2(\cdot)$  all *scalar* functions in  $H^\infty$ . This problem can be considered a robust version of the standard two-block problem

$$\inf_{Q(\cdot) \in H^\infty} \left\| \begin{bmatrix} P_1(z) + P_2(z)Q(z) \\ Q(z) \end{bmatrix} \right\|_\infty, \quad (2)$$

that arises in  $H^\infty$  full-information control, and in  $H^\infty$  estimation, which further allows one to consider possible modeling errors  $\Delta P_1(\cdot)$  and  $\Delta P_2(\cdot)$  for the nominal

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plants  $P_1(\cdot)$  and  $P_2(\cdot)$ . In problem (1) both the objective and the modeling errors are measured in the  $H^\infty$  norm. Thus  $\delta > 0$  is a measure of the modeling error allowed for in (1).

## 2 Main Result

Problem (1) is, of course, a highly nonlinear problem and satisfactory solutions to date do not exist. In this paper, we show that for “small enough” modeling errors problem (1) can be solved by doing a one-dimensional search over the solution of a certain family of weighted standard two-block problems.

Thus, consider the solution to the following weighted two-block problem:

$$f(\epsilon) \triangleq \inf_{Q(\cdot) \in H^\infty} \left\| \begin{bmatrix} \sqrt{1+\epsilon} (P_1(z) + P_2(z)Q(z)) \\ \sqrt{1+(1+\frac{1}{\epsilon})\delta^2} Q(z) \end{bmatrix} \right\|_\infty^2 + (1+\frac{1}{\epsilon})\delta^2, \quad \epsilon > 0. \quad (3)$$

The above problem can be readily solved for any value of  $\epsilon$  (say, by using Riccati-based techniques when  $P_1(\cdot)$  and  $P_2(\cdot)$  are rational), and so  $f(\epsilon)$  is easy to compute. Moreover, it can be shown that  $f(\cdot)$  is, in general, a nonconvex continuous function of  $\epsilon$ . Suppose now that we perform a one-dimensional search over  $\epsilon > 0$ , and determine

$$\epsilon^* \triangleq \inf_{\epsilon > 0} f(\epsilon). \quad (4)$$

Then we have the following result.

### Theorem 1 (Robust Two-Block Problem)

*There exists a  $\bar{\delta} > 0$ , such that for all  $\delta < \bar{\delta}$ , the solution to the robust two-block problem (1) can be found from the solution to the weighted two-block problem*

$$\inf_{Q(\cdot) \in H^\infty} \left\| \begin{bmatrix} \sqrt{1+\epsilon^*} (P_1(z) + P_2(z)Q(z)) \\ \sqrt{1+(1+\frac{1}{\epsilon^*})\delta^2} Q(z) \end{bmatrix} \right\|_\infty^2 + (1+\frac{1}{\epsilon^*})\delta^2. \quad (5)$$

The above theorem states that if the modeling error is less than  $\bar{\delta} > 0$ , then the robust two-block problem (1) can be solved using a one-dimensional search over a family of weighted two-block problems. Moreover, the solution to (1) is the same as the solution to a certain weighted two-block problem, with optimal weighting determined by  $\epsilon^*$ . This has rather interesting physical implications since it states that the modeling errors  $\Delta P_1(\cdot)$  and  $\Delta P_2(\cdot)$  can be dealt with by appropriately weighting the modeling-error-free two-block problem (2).

Of course, this result raises several issues:

- How does Theorem 1 generalize to matrix plants?
- How does Theorem 1 generalize to four-block problems?
- How large is the value of  $\bar{\delta}$  in Theorem 1?

Currently all three questions are open. To gain some insight into the third question, let us consider the robust tracking problem.

### 3 Robust Tracking

The  $H^\infty$  tracking problem corresponds to  $P_1(z) = 1$  and  $P_2(z) = -P(z)$ , so that the robust tracking problem takes the form

$$\inf_{Q(\cdot) \in H^\infty} \sup_{\|\Delta P(\cdot)\|_\infty \leq \delta} \left\| \begin{bmatrix} 1 - (P(z) + \Delta P(z)) Q(z) \\ Q(z) \end{bmatrix} \right\|_\infty. \quad (6)$$

The modeling-error-free tracking problem,

$$\inf_{Q(\cdot) \in H^\infty} \left\| \begin{bmatrix} 1 - P(z) Q(z) \\ Q(z) \end{bmatrix} \right\|_\infty \triangleq \gamma_{opt}, \quad (7)$$

has been studied in [1], where it is shown:

- If  $P(z)$  is minimum phase, then

$$\gamma_{opt} = \frac{1}{1 + \min_{\omega \in [0, 2\pi]} |P(e^{j\omega})|^2}. \quad (8)$$

- If  $P(z)$  is nonminimum phase, then

$$\gamma_{opt} = 1. \quad (9)$$

Let us now return to problem (6). Clearly, if  $P(z)$  is nonminimum phase, we can always obtain an  $H^\infty$  norm of unity in the objective cost by setting  $Q(\cdot) = 0$ . This is the same value obtained in the modeling-error-free case. Therefore, let us focus on the case where  $P(z)$  is minimum phase.

Here we will have to distinguish between two cases:

- (i)  $\delta \geq \min_{\omega \in [0, 2\pi]} |P(e^{j\omega})|^2 \triangleq p_{min}$ . In this case, there exist modeling errors for which  $P(\cdot) + \Delta P(\cdot)$  is non-minimum phase. Thus here the best choice is  $Q(\cdot) = 0$ , which results in an  $H^\infty$  norm of unity.
- (ii)  $\delta < \min_{\omega \in [0, 2\pi]} |P(e^{j\omega})|^2 \triangleq p_{min}$ . In this case,  $P(\cdot) + \Delta P(\cdot)$  is always minimum phase.

Thus, clearly the case of interest is case (ii), above. The next result shows that for this case, the optimally-weighted two-block problem *always* solves (6).

**Theorem 2 (Robust Tracking)** *Consider problem (6) and suppose that*

$$\delta < \min_{\omega \in [0, 2\pi]} |P(e^{j\omega})|^2 \triangleq p_{min}.$$

*Then the solution to problem (6) is given by the solution to the problem,*

$$\inf_{Q(\cdot) \in H^\infty} \left\| \begin{bmatrix} \sqrt{1 + \epsilon^*} (1 - P(z) Q(z)) \\ \sqrt{1 + (1 + \frac{1}{\epsilon^*}) \delta^2} Q(z) \end{bmatrix} \right\|_\infty^2, \quad (10)$$

where

$$\epsilon^* = \arg \min_{\epsilon > 0} \inf_{Q(\cdot) \in H^\infty} \left\| \begin{bmatrix} \sqrt{1 + \epsilon} (1 - P(z) Q(z)) \\ \sqrt{1 + (1 + \frac{1}{\epsilon}) \delta^2} Q(z) \end{bmatrix} \right\|_\infty^2. \quad (11)$$

In particular, when  $p_{min} < 2$ , we have

$$\epsilon^* = \frac{\delta(p_{min} - \delta)}{1 - \delta(p_{min} - \delta)}, \quad (12)$$

and the optimal  $H^\infty$  norm becomes

$$\gamma_{opt} = \frac{1}{1 + (p_{min} - \delta)^2}, \quad (13)$$

which is the same as that obtained from max-min problem:

$$\sup_{\|\Delta P(\cdot)\|_\infty \leq \delta} \inf_{Q(\cdot) \in H^\infty} \left\| \begin{bmatrix} 1 - (P(z) + \Delta P(z)) Q(z) \\ Q(z) \end{bmatrix} \right\|_\infty. \quad (14)$$

Thus the method presented here for solving robust two-block problems always works for the tracking problem. Moreover, it is interesting that for  $p_{min} < 2$  the solutions to the min-max problem (6) and the max-min problem (14) coincide. [In general we have min-max  $\geq$  max-min.]

### References

- [1] B. Hassibi and T. Kailath. Tracking with an  $H^\infty$  criterion. In *Proceedings of the 36th IEEE Conference on Decision and Control*, San Diego, CA, Dec. 1997.